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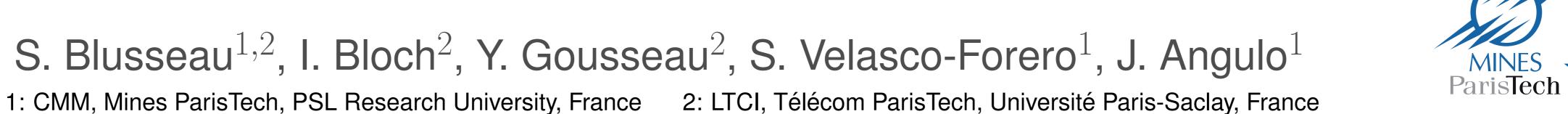
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MATHEMATICAL MORPHOLOGY ON TENSOR IMAGES FOR FIBER ENHANCEMENT

S. Blusseau^{1,2}, I. Bloch², Y. Gousseau², S. Velasco-Forero¹, J. Angulo¹





We introduce morphological methods to analyse images of positive semi-definite matrices (ellipsoids), such as diffusion or structure tensors. Morphological operators cannot be directly defined on these objects because they lack a complete lattice structure. Therefore, we extend recent developments in morphological signal processing on graphs. Graphs are constructed based on all the available information, whereas a scalar signal (anisotropy) is processed on these graphs, successfully enhancing structures such as vessels and fibers.

Basics of Mathematical Morphology

Complete lattices and morphological operators

Complete lattice A complete lattice (L, \leq) is a partially ordered set for which every subset $A \subseteq L$ has a supremum noted $\bigvee A$ and an infimum noted $\bigwedge A$. In the following, (L, \leq) and (L', \leq') denote two complete lattices.

Dilation, Erosion

- A dilation is a mapping $\delta: L \to L'$ that commutes with the supremum: for any family $(\mathbf{x}_k)_{k \in K}$ of L, $\delta(\bigvee_{k \in K} \mathbf{x}_k) = \bigvee_{k \in K}' \delta(\mathbf{x}_k)$, where K is an index set. - An *erosion* is a mapping $\varepsilon: L' \to L$ that commutes with the infimum: $\varepsilon(\bigwedge_{k \in K}' \mathbf{x}_k) = \bigwedge_{k \in K} \varepsilon(\mathbf{x}_k)$.

Opening, Closing

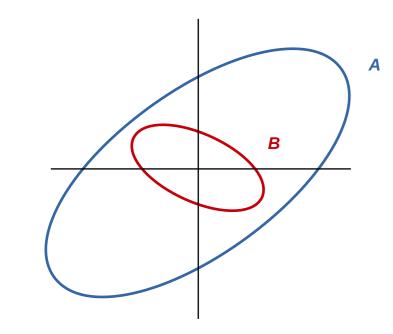
- A closing is a mapping $\phi: L \to L$ that is increasing, extensive and idempotent.
- An *opening* is a mapping $\gamma: L \to L$ that is increasing, anti-extensive and idempotent.

Adjunction A pair of mappings (ε, δ) , $\varepsilon : L' \to L$, $\delta : L \to L'$, forms an adjunction if and only if $\forall x \in L, y \in L', \delta(x) \le' y \iff x \le \varepsilon(y).$

The data

The cone of positive semi-definite matrices S_n^+

No natural lattice ordering



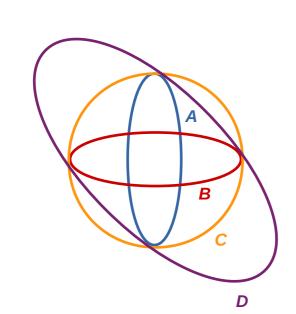


Figure 2: The Loewner ordering is defined by $B \leq A \iff A - B \in S_n^+$ (Left). The example on the right shows that this is not a lattice ordering: if it were, then C would be the supremum of A and B, but D is another upper bound of A and B and yet is not comparable with (and therefore not larger than) C.

Examples

Proposition 1. If (ε, δ) is an adjunction, then $\phi = \varepsilon \delta$ is a closing and $\gamma = \delta \varepsilon$ is an opening.

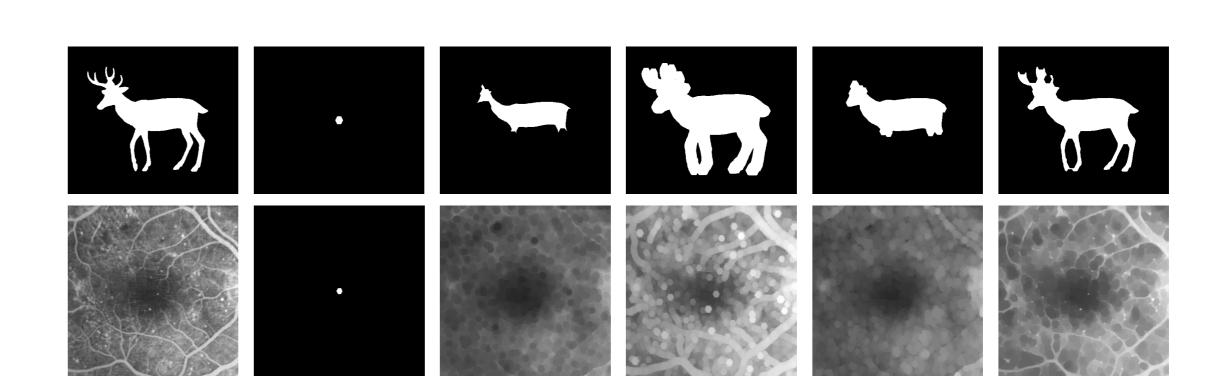


Figure 1: Example of morphological operators depending on a structuring element. (a) Input image: binary image on the first row, greyscale image on the second row. (b) Structuring element (SE) \mathcal{B} . (c) Erosion $\varepsilon_{\mathcal{B}}$. (d) Adjoint dilation $\delta_{\mathcal{B}}$. (e) Opening $\delta_{\mathcal{B}}\varepsilon_{\mathcal{B}}$. (f) Closing $\varepsilon_{\mathcal{B}}\delta_{\mathcal{B}}$.

Structure tensors for greyscale 2D images

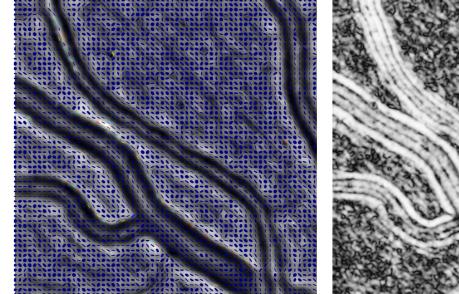
We consider a greyscale image f, whose pixels coordinates are noted (x, y).

Scale zero tensor $T_0: (x,y) \mapsto \nabla f(x,y) \cdot \nabla f(x,y)^T$

Scale $\sigma > 0$ tensor $T_{\sigma} = G_{\sigma} * T_0$ where G_{σ} is the σ -scale Gaussian kernel.

 $T_{\sigma}(x,y)$.

Anisotropy image $A_{\sigma} = \frac{\lambda_1 - \lambda_2}{\lambda_1 + \lambda_2}$, where $0 \le \lambda_2(x,y) \le \lambda_1(x,y)$ are the eigenvalues of



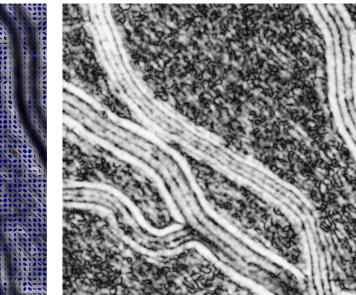


Figure 3: Image of structure tensors T_{σ} (obtained thanks to [2]) and the corresponding anisotropy image ($\sigma = 3$).

Morphological signal processing on graphs

Graph. G = (V, E), weighted and directed graph, containing n vertices.

Adjacency matrix. $W \in Mat_n(\mathbb{R} \cup \{-\infty\})$ such that for any $(i,j), 1 \leq i,j \leq n - \infty \leq w_{ij} \leq 0$, $w_{ii} = 0$, and $-\infty < w_{ij} \iff (i, j) \in E$ (called Conservative morphological weight matrix in [3]).

Lattice. We work with vectors $\mathbf{x} \in [0,1]^n$ supported by the graph G, so the considered complete lattice is $L = ([0,1]^n, \leq)$ equipped with the usual product partial ordering (Pareto ordering):

$$\mathbf{x} \leq \mathbf{y} \iff x_i \leq y_i \ \forall i \in \{1, \dots, n\}.$$

Dilations, erosions, openings. We define the dilation δ_W and its adjoint erosion ε_W on L by

$$\delta_W(x)_i = \max_{1 \le j \le n} \left(x_j + w_{ij} \right) = \max_{j \in \mathcal{N}_i} \left(x_j + w_{ij} \right), \quad \varepsilon_W(x)_i = \min_{1 \le j \le n} \left(x_j - w_{ji} \right) = \min_{j, i \in \mathcal{N}_j} \left(x_j - w_{ji} \right)$$
 (1)

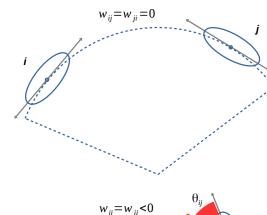
where $\mathcal{N}_i \doteq \{j \in \{1, \dots, n\}, (i, j) \in E\}$ is the set of neighbours of vertex i in G. We also consider their iterated erosions ε_W^p and dilations δ_W^p , from which we build the openings $\gamma_W^{(p)}$ for $p \in \mathbb{N}$:

$$\varepsilon_W^p = \varepsilon_W \circ \cdots \circ \varepsilon_W, \qquad \delta_W^p = \delta_W \circ \cdots \circ \delta_W, \qquad \gamma_W^{(p)} = \delta_W^p \varepsilon_W^p.$$

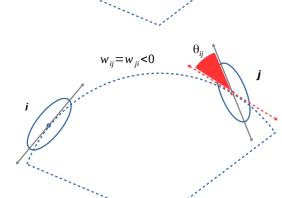
Co-circularity graph

In our application the vertices are the pixels or voxels of an image (2D or 3D), thus characterised by their coordinates and a scalar value (the anisotropy) [1].

Measure of co-circularity



Coefficient w_{ij} is computed based on the principal orientations of the ellipsoids in i and j, and their relative positions in the image. Given the angular error θ_{ij} to cocircularity (see illustration), we distinguish two cases:

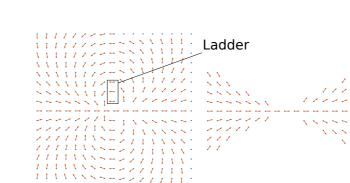


Binary case with threshold α : $w_{ij} = \begin{cases} 0 & \text{if } \cos(\theta_{ij}) \geq \cos(\alpha) \\ -\infty & \text{otherwise} \end{cases}$

Non-binary case:

 $w_{ij} = \log(\cos(\theta_{ij}))$

Additional constraints



Spatial constraint: Only pairs of vertices whose coordinates lie in a limited spatial neighbourhood can be neighbours. Denoting u_i and u_j the coordinates of pixels (or voxels) i and j, we impose $w_{ij} > -\infty \iff ||u_i - u_j||_{\infty} \le k$, for a fixed integer k. Conic constraint: Co-circularity allows ladder configurations. We avoid these by an additional constraint.

Property of the graph



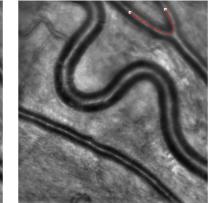




Figure 4: Maximal weight paths between pairs of vertices in a binary co-circularity graph. The weight of a path i_1, i_2, \ldots, i_q is the non-positive sum $\sum_{l=1}^{q-1} w_{i_l i_{l+1}}$. The paths with large weights are those with weights close to 0, and which therefore have the strongest contribution in the dilations and erosions defined above.

Results and discussion

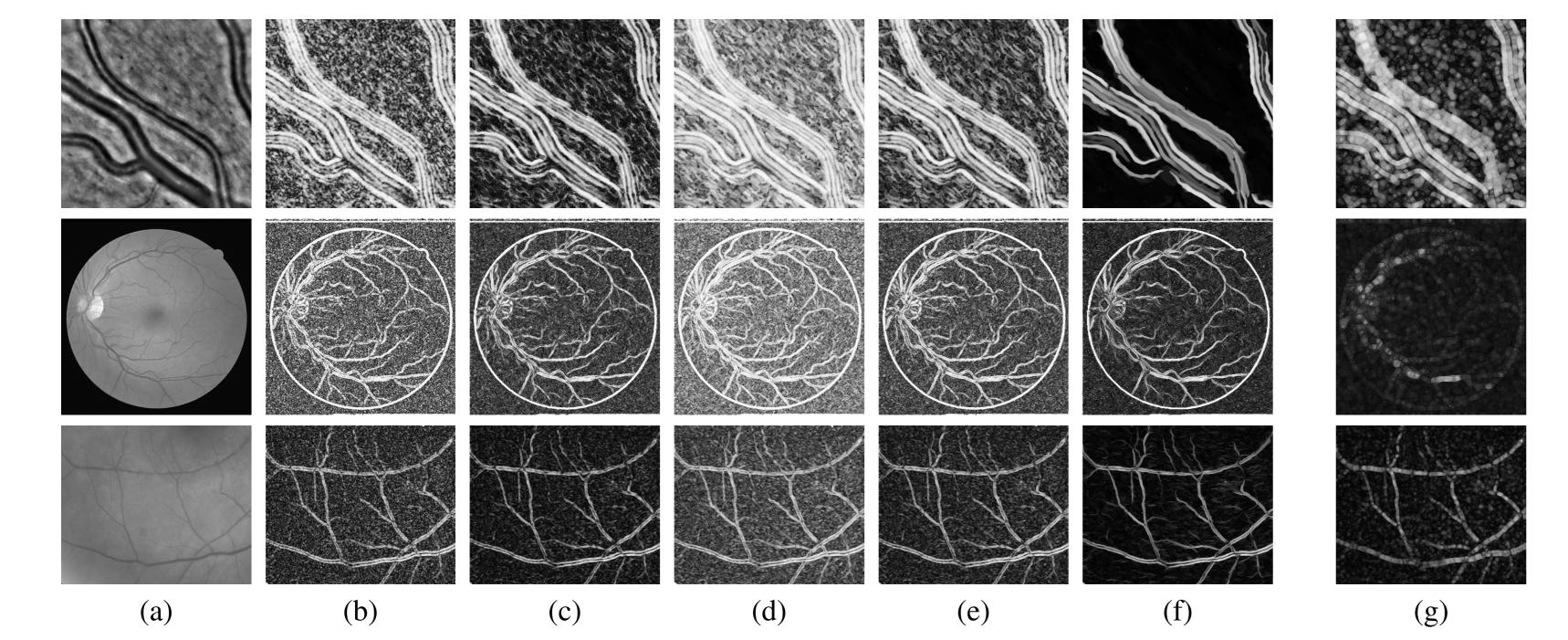


Figure 5: Results of the presented method based on co-circularity graphs (except last column). (a) Input images, (b) Anisotropy images (first and third rows: $\sigma = 3$; second row: $\sigma = 1$), (c) erosions ε_W , (d) dilations δ_W (e) openings γ_W , (f) openings $\gamma_W^{(p)}$ (first row: p = 10, second and thrid rows: p=5). Here the adjacency matrices W are binary, with angular threshold $\alpha=\frac{\pi}{6}$ and spatial neighbourhood parameter k (first row: k=7; second row: k=3, third row: k=5). (g) Results of directional openings by a family of straight lines structuring elements (classical method in MM).

Future work

- multi-scale approach or an automatic scale determination,

interesting behaviours for the processing of anisotropic structures.

- comparison with other methods, such as path openings and other morphological connected filters,

Contributions We introduced a framework that generalises mathematical morphology for sig-

nals on graphs. The operators we defined based on the co-circularity of structure tensors show

- exploration of the properties of non-binary co-circularity graphs,
- links between these operators and the max-plus algebra [1].

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